

Homework 4

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1 Problem Set 1

In this problem, we'll verify using R that SVD and Eigenvalues are related as worked out in the weekly module. Given a 3×2 matrix A

$$A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 0 & 4 \end{bmatrix}$$

1.1 write code in R to compute $X = AA^T$ and $Y = A^T A$.

```
A <- matrix(c(1,-1,2,0,3,4), nrow=2)
X <- A%*%t(A)
Y <- t(A)%*%A
list("X" = X, "Y" = Y)
```

```
## $X
##      [,1] [,2]
## [1,]   14   11
## [2,]   11   17
##
## $Y
##      [,1] [,2] [,3]
## [1,]    2    2   -1
## [2,]    2    4    6
## [3,]   -1    6   25
```

1.2 Then, compute the eigenvalues and eigenvectors of X and Y using the built-in commands in R.

```
results <- list("X" = eigen(X))
results <- c(results, "Y" = eigen(Y))
results

## $X
## $X$values
## [1] 26.601802  4.398198
##
## $X$vectors
##           [,1]      [,2]
## [1,] 0.6576043 -0.7533635
## [2,] 0.7533635  0.6576043
##
##
## $Y.values
## [1] 2.660180e+01 4.398198e+00 1.058982e-16
##
## $Y.vectors
##           [,1]      [,2]      [,3]
## [1,] -0.01856629 -0.6727903  0.7396003
## [2,]  0.25499937 -0.7184510 -0.6471502
## [3,]  0.96676296  0.1765824  0.1849001
```

1.3 Then, compute the left-singular, singular values, and right-singular vectors of A using the svd command.

```
results <- svd(A)
names(results) <- c("singular", "left-singular", "right-singular")
results

## $singular
## [1] 5.157693 2.097188
##
## $`left-singular`
##           [,1]      [,2]
## [1,] -0.6576043 -0.7533635
## [2,] -0.7533635  0.6576043
##
## $`right-singular`
##           [,1]      [,2]
## [1,]  0.01856629 -0.6727903
## [2,] -0.25499937 -0.7184510
## [3,] -0.96676296  0.1765824
```

1.4 Examine the two sets of singular vectors and show that they are indeed eigenvectors of X and Y.

```
results$`left-singular`
```

```
##           [,1]      [,2]
## [1,] -0.6576043 -0.7533635
## [2,] -0.7533635  0.6576043
```

```
X_vectors <- eigen(X)$vectors
X_vectors
```

```
##           [,1]      [,2]
## [1,]  0.6576043 -0.7533635
## [2,]  0.7533635  0.6576043
```

We can see that the two sets of vectors are indeed eigenvectors of X and Y. Also, note that the sign switch does not impact the interpretation of the eigenvectors, we can multiply by -1 and it has no further impact.

```
results$`left-singular`
```

```
##           [,1]      [,2]
## [1,] -0.6576043 -0.7533635
## [2,] -0.7533635  0.6576043
```

```
X_vectors[,1] <- (X_vectors[,1]*-1)
X_vectors
```

```
##           [,1]      [,2]
## [1,] -0.6576043 -0.7533635
## [2,] -0.7533635  0.6576043
```

```
results$`right-singular`
```

```
##           [,1]      [,2]
## [1,]  0.01856629 -0.6727903
## [2,] -0.25499937 -0.7184510
## [3,] -0.96676296  0.1765824
```

```
Y_vectors <- eigen(Y)$vectors
Y_vectors[,1] <- (Y_vectors[,1]*-1)
Y_vectors
```

```
##           [,1]      [,2]      [,3]
## [1,]  0.01856629 -0.6727903  0.7396003
## [2,] -0.25499937 -0.7184510 -0.6471502
## [3,] -0.96676296  0.1765824  0.1849001
```

2 Problem Set 2

Using the procedure outlined in section 1 of the weekly handout, write a function to compute the inverse of a well-conditioned full-rank square matrix using co-factors. In order to compute the co-factors, you may use built-in commands to compute the determinant. Your function should have the following signature:

`B = myinverse(A)`

where `A` is a matrix and `B` is its inverse and $A \times B = I$. The off-diagonal elements of I should be close to zero, if not zero. Likewise, the diagonal elements should be close to 1, if not 1. Small numerical precision errors are acceptable but the function `myinverse` should be correct and must use co-factors and determinant of `A` to compute the inverse.

```
myinverse <- function(M){
  results <- list()
  results$original <- M
  #save original to compare
  cofactors <- matrix(nrow = nrow(M), ncol = ncol(M))
  # create empty matrix to store cofactors
  for(i in 1:nrow(M)){ #loop over rows
    for (j in 1:ncol(M)) # loop over columns
      cofactors[i,j] <- ((-1)^(i + j)*det(M[-i, -j]))
      #for row, column; sign * determinate of submatrix
    }
  results$inverse <- t(cofactors)/det(M)
  #transpose of cofactors matrix / determinate of original matrix
  return(results)
}
```

```
M <- matrix(c(1:7,12,20), nrow = 3)
myinverse(M)
```

```
## $original
##      [,1] [,2] [,3]
## [1,]    1    4    7
## [2,]    2    5   12
## [3,]    3    6   20
##
## $inverse
##      [,1]      [,2]      [,3]
## [1,] -3.111111  4.222222 -1.444444
## [2,]  0.444444  0.111111 -0.222222
## [3,]  0.333333 -0.666667  0.333333
```